Kinematically irreversible acinar flow: a departure from classical dispersive aerosol transport theories

F. S. HENRY, J. P. BUTLER, AND A. TSUDA

1School of Engineering, City University, London EC1V 0HB, United Kingdom; and 2Physiology Program, Harvard School of Public Health, Boston, Massachusetts 02115

Received 23 April 2001; accepted in final form 22 October 2001


The objectives of the studies reported here are, through numerical simulations of bolus experiments, to provide key data showing that the behavior of particles in a rhythmically expanding, multiply alveolated lung is chaotic. The multiscale nature of this chaotic mixing is radically different from diffusive mixing. Research in fluid mechanics has shown that chaotic mixing can occur even in a viscous flow. In chaotic acinar flow, a tracer bolus undergoes cyclic stretch-and-fold deformation, resulting in the induction of finer and finer scales in tracer profile with repeated breaths. This process soon reaches a critical moment at which the lateral distance between adjacent tracer striations becomes comparable to the diffusion distance. A burst of mixing occurs at this moment, and mixing is quickly completed.

The costs of publication of this article were defrayed in part by the payment of page charges. The article must therefore be hereby marked “advertisement” in accordance with 18 U.S.C. Section 1734 solely to indicate this fact.

Address for reprint requests and other correspondence: A. Tsuda, Physiology Program, Harvard School of Public Health, Huntington Ave., Boston, MA 02115 (E-mail: atsuda@hsph.harvard.edu).

http://www.jap.org 8750-7587/02 $5.00 Copyright © 2002 the American Physiological Society 835
duct flow, with a saddle point and associated vortexes in each air pocket, does not satisfy the fundamental assumptions of any dispersion theory and that the shape of a tracer bolus evolves to a stretch-and-fold fractallike pattern, similar to those found in flow visualization experiments in rat lungs (Tsuda A, Butler JP, and Rogers RA, unpublished observations). The results suggest that 1) kinematic irreversibility is the origin of aerosol transport, 2) axial transport cannot be characterizable by an effective diffusivity, and 3) fractal trajectories can occur in most of the alveoli in the acinar tree. The alternative mechanism of aerosol transport that we propose here may, in fact, be the dominant mechanism determining deposition of submicrometer particles deep in the lung.

METHODS

In a previous investigation (40), we used the single-alveolus model to explore the basic physics operating in a viscous flow subjected to cyclic alveolar wall motion. The alveolus model used in the present investigation is also axisymmetric but comprises a central circular channel around which are placed nine tori, equispaced in the axial direction (Fig. 1A). Details of a typical cell are given in Fig. 2. The duct and alveolar walls move in a perfectly kinematically reversible, simple sinusoidal manner with a specific volume excursion \( C \) of 25\% \( |C = (V_{\text{max}} - V_{\text{min}})/V_{\text{min}}| \), where \( V_{\text{max}} \) and \( V_{\text{min}} \) are the maximum and minimum volumes of the model, respectively] and a cycle period \( T \) of 3 s. These correspond roughly to typical tidal ventilation and respiratory period in human. Any length scale \( L \) of the model changes as \( L(t) = L(1 + K \sin(nt)) \), where \( L \) is the mean \( L \) value, \( t \) is time, \( n = 2\pi/T \), and \( K = (1 + C)/C \). A smooth-walled, closed-end duct is fitted to the distal end of the model. The closed-end duct is used to approximate the airway distal of the site of interest up to the terminal alveoli, and it is also used to control the Reynolds number of the flow in the model.

Fig. 1. Schematics of moving walled models. A: 9-cell alveolated duct model. B: isolated cell alveolated duct model. C: nonalveolated smooth-wall model.

Fig. 2. Expanded view of a typical alveolar cell. \( R_D (= 250 \mu m) \), duct radius; \( R_A (= 200 \mu m) \), alveolar radius; \( L_A (= 346 \mu m) \), alveolar opening length; \( L_C (= 413 \mu m) \), alveolar cell length; \( Q_D \), ductal volume flow rate; \( Q_A \), alveolar volume flow rate; \( \gamma (= 60^\circ) \), alveolar half-opening angle; \( C_L \), duct center line. Note that the alveolar corners were rounded with the corner radius = 0.02 \( R_D \).
The walls of this duct also move in the same sinusoidal manner as the main alveolus model. The mean length of this duct, $L_D$, controls the bulk velocity $U$ of the fluid entering the model, i.e., $U = Q/\pi R_D^2 = 3nKLD \cos(\epsilon)$, where $Q$ is the volume flow rate and $R_D$ is the duct radius. The root mean square (RMS) Reynolds number $Re_{RMS} = URMS/R_D\nu$, where $URMS$ is the RMS $U$ and equals $U_{max}/\sqrt{2}$, where $U_{max}$ is maximum $U$, $R_D$ is the mean $R_D$, and $\nu$ is the kinematic viscosity. $Re_{RMS}$ ranges from 0.006 to 0.728 in the alveolated region of the model. Using a closed-end duct has the added advantage of avoiding the difficulty of defining appropriate boundary conditions at the downstream boundary of the alveolated section. Typically, $L_D/R_D > 100$, and, hence, it can be assumed that the end of the closed-end duct was sufficiently far from the downstream boundary of the alveolated section so as not to affect the flow across this boundary adversely. As a further refinement to the original model, the sharp corner at the intersection between the alveolus and the duct is replaced by a more natural circular section. For comparative purposes, two other cases are also simulated: one for an isolated alveolus model (Fig. 1B), similar to that used in Tsuda et al. (40), and the other is the flow in a nonalveolated, rhythmically expanding straight tube (Fig. 1C).

The flow field is defined by the full, incompressible Navier-Stokes equations, which are solved numerically on a multiblock, body-fitted moving grid using the finite volume code CFX-4 (CFDS, AEA Technology, Harwell, UK). This general-purpose, pressure-correction code offers a variety of discretization schemes and solution techniques. In these calculations, central differencing is used to model the convection terms, and the implicit backward Euler method is used to advance the solution in time. A combination of the SIMPLEC pressure-correction method (44) and the Rhie-Chow (28) algorithm to eliminate pressure oscillations on the collocated grid (12) is used in the formulation of the discrete equations. Stone’s method (33) is used to solve the discrete velocity equations, and the method of preconditioned conjugate gradients (see, for example, Ref. 22) is used to solve the discrete pressure correction equation. At the inlet, a constant-pressure boundary is defined. The value of the pressure on this boundary is set arbitrarily to zero, and, as for incompressible flow, only the pressure gradient is of importance. The no-slip condition is enforced on all solid surfaces, which, in the case of moving walls, means that the fluid matches the wall velocity at the fluid-wall interface. Tests were carried out to ensure that the solutions are grid independent and converged. For example, increasing the number of grid cells by 30% above that which was eventually used produces an increase of only 0.16% in the predicted maximum velocity. The final 73-block grid had a total of 38,349 active cells. With the use of a measure of error due to using backward Euler time stepping, suggested by Roache (29), a time step of $T/240$ is found to give sufficiently accurate results in that further reductions in the size of the time step used does not produce any significant reduction in the error. These simulations are computationally intensive, with one breathing cycle taking ~350 h on a Sun Ultra 10.

Particle trajectories are calculated in all three models using a special-purpose tracking routine. This routine reads a full cycle of flow field and grid data produced by CFX-4 and uses this data repeatedly, cycle after cycle, and a predictor and corrector method to track individual particles (fluid elements) over as many cycles as required. The time step used in the particle tracking routine is set independently of that for the flow-field solution. Before the particle track is advanced to the next time, the time step is recalculated, using local flow conditions, to ensure that the particle does not step out of the solution domain. At each particle-track step, a grid and flow field are created from the CFX-4 data using bicubic interpolation in space (27) and linear interpolation in time. For numerical efficiency, the routine solves the particle tracks in a stationary computational space, and, at each time step, the particle position is mapped back to physical space before it is written to an output file. The maximum error in the predicted particle position is estimated to be 0.2% of the distance traveled by the particle, and, in most cases, the error is considerably smaller than this.

RESULTS

Flow patterns. Solving the velocity field of the carrier gas on a moving grid over the physiologically relevant range of flow parameters ($Re_{RMS} < 1$) in a rhythmically expanding and contracting, multiply alveolated duct, we often detected the presence of slowly rotating recirculation in each alveolus (see Fig. 3, A and B). The size of the recirculation flow depends on the ratio between the alveolar flow ($Q_a$) (i.e., flow produced by the volume change of the alveolus) and the volumetric ductal flow ($Q_d$) ($Q_a/Q_d$). Similar to our previous study (40), we found that the smaller the $Q_a/Q_d$, the larger the alveolar recirculation. When $Q_a/Q_d$ is larger than ~0.1, however, the alveolar flow is largely radial without recirculation (Fig. 3C). Because, under normal breathing conditions or during moderate exercise, the value of $Q_a/Q_d$ is usually <0.05 in the majority of alveoli along the acinar tree (from the respiratory bronchioles to the last few generations) (37), we expect that most of the alveoli are likely to possess recirculation in their flow field. Importantly, the presence of alveolar recirculation in the cyclically expanding alveoli is topologically associated with the existence of a stagnation saddle point in each alveolar flow field (40), implying that chaotic mixing could originate in most of the alveoli (see DISCUSSION).

Interalveolar kinematic mixing. To demonstrate the effects of a series of saddle points on fluid flow irreversibility, the motion of massless particles was tracked over one ventilation cycle in a nine-cell alveolar model for three different ranges of $Q_a/Q_d$ (0.0050 < $Q_a/Q_d$ < 0.0053, 0.040 < $Q_a/Q_d$ < 0.069, and 0.081 < $Q_a/Q_d$ < 0.577 shown in Fig. 4, A, B, and C, respectively). The particles were initially placed on radial lines across the duct midway between alveoli at three different initial axial locations (shown as three different colors, brown, green, and pink, in Fig. 4). As soon as inspiration begins, particles near the center line convect distally, proportionally to the bulk mean velocities (see $t/T = 0.25$ in Fig. 4). The lines of particles quickly approach each other and comigrate along the central channel, particularly in the cases of smaller $Q_a/Q_d$ (Fig. 4, A or B). The particles that were located initially near the alveolar opening enter the alveolus (Fig. 4). The depth of particle penetration into the alveolus during inspiration depends on the size of alveolar recirculation. When the alveolar recirculation is large (Fig. 4A), the particles penetrate deep into the alveoli, and the par-
that remained in the central channel during inspiration virtually retrace their inspiratory paths during expiration and thus arrive back very close to their original starting position. Comparing these three types of particle behavior, we notice that particles that were associated with alveolar recirculation (even for a short period of time) followed irreversible trajectories. When the alveolar recirculation is present but small (Fig. 4B), no particles are trapped in the alveolus, suggesting that the effects of small alveolar recirculation are insufficient to cause particle trapping. However, particles initially located near the side walls (which eventually became associated with alveolar recirculation) still display irreversible motion (type II). Particles that start near the channel center line come back to their original starting position (type III). In the complete absence of alveolar recirculation (Fig. 4C), the motion of all particles, even those that traveled inside the alveolus, is reversible.

For the purpose of comparison, the following two additional cases were also conducted: massless particles were tracked in a single-cell model and in an expandable straight-tube model (data not shown). Whereas the particles in the single-alveolus model, similar to the nine-cell alveolar model, exhibit some irreversibility, especially near the walls (type II), the particles in the expandable straight tube are reversible as predicted (42).

**Motion of interface between inhaled tracer fluid and the host alveolar residual fluid.** Regarding massless tracer particles as marked fluid elements, the interface between one fluid and another can be approximated by chords connecting initially adjacent particles (Fig. 5A). In our studies, this piecewise linear approximation to the interface can represent the front of incoming tidal gas or a tracer bolus facing the host alveolar residual gas. As we have demonstrated in Fig. 4, A and B, the kinematically irreversible deformation of the tracer interface seems to be due to its association with alveolar recirculation flow over the respiratory cycle. Thus characteristics (e.g., size and strength) of the alveolar recirculation may be major determinants in these processes. Our studies described here focus on the situation that presumably occurs deep in the acinus, where the alveolar flow exhibits a medium- to small-sized recirculation (Fig. 3B). Larger alveolar recirculation flow (Fig. 3A) is likely to occur near the entrance of the acinus (e.g., respiratory bronchioles), and that case is discussed elsewhere (38).

To understand the temporal evolution of interface deformation (approximated as described above) in the case of flows with medium-sized alveolar recirculation, we monitored the motion of a tracer bolus for three cycles. In these simulations, we systematically increased the number of massless particles $P$ ($P = 2^j$; where $j = 3, 4, 5, \ldots, 13$). For each $P$, the initial radial distribution of particles was adjusted in such a way that each particle represented equal cross-sectional annular area. Over each cycle, the interface progressively deforms into “fingerlike” protrusions located especially near the duct walls (Fig. 5B). Each protrusion
Fig. 4. Interalveolar mixing. Massless (fluid) particles were tracked, starting from 3 different initial axial locations (shown as 3 different colors) with 3 different flow conditions. A: 0.0050 < $Q_A/Q_D$, 0.728 > Re$_{RMS}$ > 0.690. B: 0.040 < $Q_A/Q_D$, 0.092 > Re$_{RMS}$ > 0.053. C: 0.081 < $Q_A/Q_D$, 0.577 < Re$_{RMS}$ > 0.006. Five different time points in each simulation were shown: time ($t$) / cycle period ($T$) / = 0 (initial), $t/T = 0.25$ (peak inspiration), $t/T = 0.5$ (end inspiration), $t/T = 0.75$ (peak expiration), and $t/T = 1.0$ (end expiration). Note that, in A and B at end inspiration, many particles are distal of the alveolated portion of the model and hence are not visible in the views given.
consists of complex stretched-and-folded patterns (Fig. 5B, inset 1), and, moreover, as the scale becomes finer, similar and finer stretched-and-folded patterns are revealed (Fig. 5B, inset 2). This suggests that the observed patterns might be qualitatively self-similar over a wide range of length scales. Such self-similarity is characteristic of fractal geometry observed in many chaotic systems (32).

To quantify the extent of stretch-and-fold flow irreversibility and especially to test whether axial spreading could be described by an effective diffusivity, we analyzed the shape of the tracer in axial and lateral directions separately. To characterize the axial phenomena, the axial variance ($\sigma^2$) of distribution of particles was computed and plotted vs. cycle number ($N$) as a family in the number of the particles ($P$) employed (Fig. 6). The results show that $\sigma^2$ is independent of $P$ and grows exponentially with increasing $N$ [$\sigma^2 = 0.0474(e^{0.457N} - 1)$]. It is important to note that $\sigma^2$ does not increase linearly with $N$; this observation is fundamentally inconsistent with the predictions of the classical dispersion theory (discussed below).

To characterize the lateral phenomena, we examined the extent of tracer stretching [$\Delta L/L_o = (L - L_o)/L_o$, where $L_o = \sum_{k=1}^{P}|X_{k+1} - X_k|$; $X_k$ is the $k$th particle position] is the initial length of the tracer, and $L = \sum_{k=1}^{P}|X_{k+1} - X_k|$; $X_k$ is the $k$th particle’s mapped
position) is the tracer length at end expiration and the pattern of the tracer. $\Delta L/L_o$ was plotted vs. $N$ as a family in $P$ (Fig. 7). The results show that the length of the tracer dramatically increases with increasing $N$ and that this increase in $\Delta L/L_o$ is strongly dependent on the $P$ used to approximate the interface. (Note that this behavior is different from the axial phenomena in which the growth of $\sigma^2$ is essentially independent of the $P$.) Furthermore, the tracer length grows exponentially with $N$ if a large $P$ ($P > 4,096$) is used for simulation. Because the reciprocal of $P$ is a parameter related to scale resolution in our simulation, the fact that the increase in $\Delta L/L_o$ with $N$ strongly depends on $P$ suggests that the resulting pattern of the tracer is fractal. To test this possibility, we examined the spatial pattern of all of the particles ($\sim 16,000$) used for the simulation described above at the end of every cycle by employing the fractal analysis method of box-counting technique popularized by Glenny (15). Briefly, the test section of interest (usually the most particle-dense area) (Fig. 8, inset) was covered by $M$ square boxes, each with an edge length $E$. Denoting the particle concentration in the $i$th box by $\mu_i$ and the overall mean concentration by $\bar{\mu}$, we computed the standard deviation $[SD = \sqrt{\sum (\mu_i - \bar{\mu})^2/M}]$ and coefficient of variation, $CV = SD/\bar{\mu}$ of the set of concentrations $\mu_i$. This procedure was repeated for box sizes spanning more than four orders of magnitude for the field of interest. A log-log plot of $CV$ vs. $E^2$ reveals a linear relationship with slopes of $-0.21$, $-0.25$, and $-0.25$ for $N = 1$, 2, and 3, respectively (Fig. 8B). This indicates a power law relationship between the tracer particle distribution and the resolution of the analysis and that, therefore, the tracer pattern resulting from kinematically

![Graph showing tracer stretching ($\Delta L/L_o$, where $L$ is length and $L_o$ is initial length) at end expiration plotted vs. $N$ as a family in $P$. $\bullet$, $P = 512$; $\triangle$, $P = 1,024$; $\blacksquare$, $P = 2,048$; $\bullet$, $P = 4,096$; $\bullet$, $P = 8,192$. Flow conditions in the 9-cell model: $Q_A/Q_D < 0.069$, $0.092 > \text{Re}_{\text{rms}} > 0.053$.](image1)

![Log-log plot of coefficient of variation (CV) vs. edge length ($E^2$) with a slope of $-0.2$ shows that the pattern of particle distribution is fractal with a fractal dimension $D \approx 1.2$. Flow conditions in the 9-cell model: $Q_A/Q_D < 0.069$, $0.092 > \text{Re}_{\text{rms}} > 0.053$.](image2)
irreversible acinar fluid mechanics may indeed be fractal with a fractal dimension $D = 1.2$ ($D = 1 - \text{slope}$; Ref. 3). Remarkably, this dimension is close to that found in our animal experiments (unpublished observations, also see Discussion).

**DISCUSSION**

The principal findings of this study are that 1) chaotic fluid motion occurring in a rhythmically expanding and contracting, multiply alveolated duct induces substantial kinematic irreversibility in the acinus, even under low-Reynolds number flow conditions, and 2) mixing due to this kinematic irreversibility is fundamentally different from the mixing described in dispersive processes.\(^1\) A tracer (bolus) subjected to the chaotic flow field is deformed both axially and laterally. The $\sigma^2$ of particle distribution increases exponentially, rather than linearly, with increasing cycle time; thus axial bolus spreading does not obey the basic rules described in classical diffusive transport theories (35). The cycle-by-cycle evolution of lateral particle distribution is even more complex. The tracer forms fingerlike protrusions, even after one cycle. The tracer length exponentially increases as $N$ increased, forming characteristic stretch-and-fold fractal-like patterns.

**Chaotic mixing in the pulmonary acinus.** Viscous flow has been considered kinematically reversible if the boundary motion is reversible (36, 45). In the mid-1980s, however, there was a breakthrough discovery in fluid mechanics, namely, that even Stokes flow can be kinematically irreversible if the structure of the flow is chaotic (2, 25). In the last several years, applying this new concept to respiratory fluid mechanics, our laboratory has been studying the role of chaotic flow phenomena in the experimentally observed, yet theoretically unexplained, convective mixing occurring in the lung periphery (5, 16, 37, 40). In our laboratory's previous numerical study (16, 40), we reported that chaotic flow and chaotic mixing can occur in the alveolated duct because of its peculiar geometry and time-dependent motion associated with tidal breathing. We found that acinar flow was often slowly rotating in the alveolar air pocket, and the velocity field near the alveolar opening was complex with a stagnation saddle point typical of chaotic flow structure. Performing Lagrangian fluid particle tracking, we further demonstrated that, in such a flow structure, the motion of fluid, $x(t)$, could be highly complex, irreversible, and unpredictable even though it was governed by simple deterministic equations $x(t) = \int v(x, t) \, dt$, $x(0) = x_0$, where $v(x, t)$ denotes Eulerian velocity field.

Our initial, numerical investigations performed in a simplified system with an isolated, single alveolus were aimed at discovering and understanding the basic physics operating in a rhythmically expanding alveolar flow (16, 40). These studies, however, did not address the cumulative effects of multiple alveoli (i.e., a series of saddle points) on the fate of inhaled aerosols. There are roughly 300 million alveoli in the human lung (∼10,000 alveoli in each of ∼30,000 acini) (17). This means that, in each acinus, the incoming tidal air may sample roughly 200 alveoli along a longitudinal pathway, from the entrance of an acinus to the terminal alveolar sac (here, we assume that alveoli are uniformly spaced in nine intra-acinar airway generations). As demonstrated in Figs. 3 and 4, a series of saddle points and associated vortexes in each air pocket, generated in a cyclically expanding, multiply alveolated duct, make the supposedly reversible low-Reynolds number acinar flow highly irreversible and can cause substantial interalveolar convective mixing.

In the context of flow irreversibilities associated with multiple saddle points, it is important to recognize that the estimates that we obtain in this work may significantly underestimate the importance of convective mixing. In particular, we have explicitly assumed and, therefore, constrained the flow field to be axisymmetric, which is to say that all azimuthal velocity components (i.e., the component in the direction perpendicular to the $r$-$z$ plane) are zero. This would imply that the presence of a saddle point in the longitudinal section would be associated with a saddle “line” or “circle” around the ductal axis. Such a flow structure is not only highly unlikely but also would be expected to be unstable and to break into a separate sequence of saddle points. Any such failure to preserve axial symmetry would thus enhance whatever convective mixing is already associated with the axisymmetric alveolated geometry.

**Fingerlike protrusion.** In this study, a line of massless particles was introduced in the alveolated duct to represent a fluid-fluid interface (Figs. 4 and 5A). The behavior of this interface, such as its reversibility and irreversibility, changes in its shape and size and contains crucial information for understanding the mechanism of mixing between inhaled particles and alveolar residual gas. By tracking the motion of the tracer particles, we have followed the motion of this interface over several cycles. Because the boundary of this line, shown as the point $Q$ in Fig. 5A, is stationary on the wall because of the no-slip condition, the line expands when the alveolar walls expand during inspiration, and the line also tends to contract when the walls contract during expiration. However, the reversibility of this process depended on the nature of the flow fields sampled during expansion and contraction. The segments of the line near the channel center line are enormously stretched axially (see Fig. 4) and sample mostly reversible Poiseuille-like ductal flow fields (Fig. 3). Consequently, the interfacial line near the center line also shows approximately reversible behavior (Fig. 5B). By contrast, the segments of the lines near the walls (e.g., segments $s_1$, $s_2$, and $s_3$ in Fig. 5A) sample a series of irreversible alveolar flow fields (e.g., Alv-1, Alv-2, and Alv-3, respectively, in Fig. 5A) and, consequently, do not return to their original positions (Fig. 5B). Each of these line segments, separated by points.

\(^1\)It is unlikely that the results on kinematic irreversibility in this paper are important to ventilation-perfusion matching, but a quantitative assessment of this remains open.
that sample approximately reversible flow fields between alveoli (e.g., R1, R2, and R3) basically form one finger after a cycle (Fig. 5B). The number of “fingers,” therefore, roughly matches the number of alveoli distal to the initial position. Although, in the present computational model, the number of alveoli that the tracer samples is limited to six (due to computational constraints), in a real acinus, the inhaled bolus is expected to encounter a larger number of alveoli (~200) along the acinar longitudinal pathway. This implies that the acinar airways are likely to be filled with many longitudinal fingers after a few breathing cycles. This prediction has been confirmed experimentally in our laboratory’s recent flow visualization studies performed in rat acini (39).

The size of each finger depended on its resident time in alveolar recirculation. The line segments that were closer to the side walls (e.g., s1 and s2) spent more time in recirculation (Alv-1 and Alv-2, respectively) and produced longer fingers. As the N increased, the tracer repeatedly encountered alveolar recirculation; the number of fingers rapidly increased, and they propagated toward the channel center line. This global evolution of the tracer pattern (i.e., rapid increase in the number of fingers, cycle-by-cycle lateral propagation), together with progressively finer scale tracer striations (discussed below in detail), are important observations because they indicate a substantial net enhancement of lateral particle transport for deposition.

Axial phenomena—a departure from the conventional dispersion theory. In current theories, aerosol transport in the pulmonary acinus is described as a dispersive (diffusion-like) process, and the mixing phenomena are couched in the language of a $D_{\text{eff}}$. The most important feature of this approach is that, in any process that is dispersive in the sense that it can be characterized by an effective diffusivity, the variance of a bolus asymptotically increases linearly in time (or $N$). Phrased differently, the variance is an additive function over time. The reduction of experimental data through the use of some $D_{\text{eff}}$ has thus been the framework by which many aerosol studies have been conducted (e.g., Ref. 30). For instance, bolus spreading in the tracheobronchial tree is commonly characterized by the difference between the inhaled and exhaled variances (proportional to $H^2$, where $H$ is the bolus width at half height), given by $\sqrt{H_{\text{exp}}^2 - H_{\text{insp}}^2}$, where $H_{\text{exp}}$ is expiratory $H$ and $H_{\text{insp}}$ is inspiratory $H$ (43). This numerical maneuver is clearly based on the underlying assumption of additivity of variances.

In sharp contrast to these ideas, the result of our numerical experiment does not obey this basic rule (Fig. 6). The $\sigma^2$ of bolus particle distribution grows faster than linearly in time, in contrast to the linear growth predicted by all theories that are characterized by a $D_{\text{eff}}$. It is important to emphasize here that mixing of kinematic origin (e.g., chaotic flow in the acinus) has a fundamentally different nature from mixing described in dispersive mechanisms; thus it cannot be described by any $D_{\text{eff}}$.

Lateral phenomena (fractal patterns). The results of our study show that complex flow phenomena can occur in the lateral direction. Because of alveolar flow irreversibility, the tracer (i.e., a series of line segments, s1, s2, etc., in Fig. 5A) deforms at every cycle, forming fine characteristic stretch-and-fold striation patterns (see Fig. 5B, insets), on top of the global fingerlike protrusion (Fig. 5B). Consequently, an enormously large lateral diffusion surface (which is represented as a stretched tracer line in our study) evolves over every cycle, suggesting a substantial enhancement of lateral particle transport and subsequent deposition on the acinar walls. Interestingly, we found that exact estimation of tracer length $L$ was not possible because $L$ was strongly dependent on the number of particles forming the tracer (Fig. 7). At sufficiently fine scales (i.e., when the number of test particles used is sufficiently large), the apparent $L$ of the tracer increases exponentially with increasing $N$.

The analysis of potentially fractal patterns by the well-known method of box counting (32) was employed. The analysis shows that the distribution of tracer particles evolve in a fractal-like manner, with $D_{\text{fractal}} \approx 1.2$. The fact that the tracer pattern exhibits fractal characteristics is not entirely surprising, because the origin of particle irreversibility is due to chaotic alveolar flow, and deterministic chaos often manifests a fractal geometry (23, 32). On the other hand, it is important and remarkable that the fingerlike protrusions (i.e., global pattern) found in these simulations are strikingly similar to those found in the longitudinal airway section of our experiments performed with rat lungs (39). Moreover, our initial finding that $D_{\text{fractal}} \approx 1.2$ obtained in finer scale stretch-and-fold striations in the present numerical study is close to $D_{\text{fractal}} \approx 1.1$ found in those rat experiments (unpublished observations) is encouraging. Although further detailed analyses will be necessary to determine the dependence of the fractal dimension (or indeed if the pattern remains fractal) on model parameters, these similarities suggest that our numerical simulation, although based on highly idealized assumptions, does, in fact, capture the essential features of the underlying mixing mechanism, namely, that low-Reynolds number chaotic flow in the acinus determines particle transport in the lung.

Physiological origins of mixing. Our laboratory has recently proposed two possible origins of “stretch-and-fold” kinematics in the pulmonary acinus: first, that...
induced by a small departure (asynchrony) from kinematically reversible motion of alveolar walls (16, 41), and, second, that due to the presence of saddle points associated with alveolar recirculation flows in the acinar flow field (40). With respect to the first mechanism, Miki et al. (24) reported the presence of a small but consistent geometric hysteresis (i.e., temporal asynchrony) in lung expansion during normal tidal ventilation in live rabbits. By matching this degree of geometric hysteresis, we have generated physiologically realistic asynchrony in physical models (16), which demonstrated that geometrical hysteresis, even if small, can produce stretch-and-fold patterns and, consequently, induce substantial acinar flow irreversibility. In a related work, Smaldone et al. (31), using gravitational sedimentation of aerosol particles to estimate mean linear intercepts, showed significant geometric hysteresis in excised lungs with large-volume excursions from minimal volume and interpreted their data in terms of respiratory unit recruitment and derecruitment. Those studies, however, being restricted to single-volume histories, did not address the issue of mixing during cyclic ventilation and so are not strictly comparable to the present work. Finally, flow/volume hysteresis (different flow magnitudes at isovolume points on inspiration and expiration) can lead to differences in the velocity profiles in the central airways; this, in turn, can cause a difference in aerosol deposition between inspiration and expiration (4).

In contrast to these studies, which focus primarily on differing geometric features between inspiration and expiration, in this paper we investigate the second mechanism mentioned above. We believe that the presence of saddle points may be an equally fundamental mixing mechanism responsible for convective mixing in the acinus. Note that such saddle point singularities do not exist in rigid wall models of acinar flow but, through their association with alveolar recirculation, are necessarily linked to the cyclic motion of the alveolar walls. To distinguish clearly this mechanism from the former one (caused by geometric hysteresis), we used only kinematically reversible wall motion in these studies. The present paper extends our previous work (40), which was restricted to a saddle point in a single alveolar space, by considering the effect of multiple saddle points in a multiply alveolated channel. In practice, we believe that both mechanisms, asynchrony and saddle points, coexist in the pulmonary acinus, mutually enhancing each other in producing stretch-and-fold mixing.

Significance and conclusions. It is well established that exposure to environmental aerosol pollutants is associated with health risks, ranging from mild to life threatening (47). The exposure and pathophysiological consequences are clearly linked by two independent and separate causal pathways: the exposure-dose relationship (i.e., given an aerosol concentration in the ambient air, what is the actual dose delivered to the lung) and the dose-response relationship (i.e., for a given deposited dose or burden, what is the biological consequence). Despite the large literature on exposure assessment methodologies as well as on the pathophysiological consequences of short- and long-term exposures, there is much less known about the mechanisms contributing to the first of these links, and much deposition and mixing data are inconsistent with previous theories of mixing deep in the lung. It is important, therefore, to identify new potential mechanisms that may dominate aerosol transport, even when boundary motion is approximately reversible. We argue in this paper that chaotic mixing is such a candidate. We have shown, through realistic numerical simulation of the low-Reynolds number alveolated duct flow (and by comparison with experimental results in rat lungs; unpublished observations), that the peculiar geometry of the alveolated duct structure within the pulmonary acinus and its cyclic motion during breathing can give rise to 1) a chaotic type of mixing associated with the presence of saddle points, 2) slow recirculatory flow within the alveoli, and 3) stretching and folding of stream surfaces. These, in turn, can significantly increase mixing, especially laterally, and will also contribute to an increasing $r^2$ (which increases faster than linearly with breath number, an observation that is inconsistent with any dispersal mechanism that can be characterized by an effective axial diffusivity). We suggest that chaotic mixing may be the dominant mechanism of aerosol transport and deposition deep in the lung.

This study was supported by National Heart, Lung, and Blood Institute Grants HL-47428 and HL-54885 and, in part, by Environmental Protection Agency Research Award R827353.

REFERENCES